A Core-Attachment based Method to Detect Protein Complexes in PPI Networks

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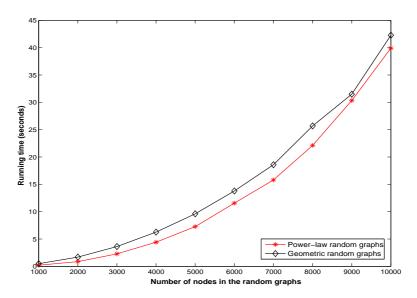
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Additional File 1

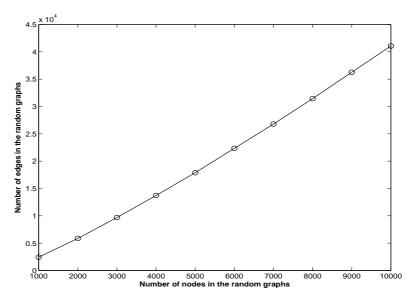
Figure 1 shows the running time of our COACH method over two kinds of random graphs (i.e., Power-law random graphs and Geometric random graphs) on the workstation with 3.4GHz Dual Core processors and 3GB RAM. In power-law graphs, the degree distribution follows a power-law, i.e., $P(k) \sim k^{-\alpha}$, where P(k) is the probability of a nodes with a degree of k and $\alpha > 0$. Geometric random graphs are constructed by dropping n points randomly into the unit square and adding edges between two nodes with distance less than a pre-defined threshold.

In our experiments, after a power-law graph is randomized, a geometric random graph will be generated with the same number of nodes and edges as the power-law graph. Given a fixed number of nodes, we generated 50 pairs of power-law graphs and geometric random graphs and then calculated the average running time of COACH over them. Figure 1 thus demonstrates that our COACH method is efficient in large-scale graphs.

The basic information of these random graphs, i.e., the average number of edges, is shown in figure 2. We also note that the average degrees $(2 \times \#Edges/\#Nodes)$ of the random graphs slightly increase as the number of nodes increases.



 ${\bf Fig.\,1.}$ The running time of our COACH on the random graphs.



 ${\bf Fig.\,2.}$ The average number of edges in the random graphs in our experiments.